

Universal features of exit probability in opinion dynamics models with domain size dependent dynamics

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Abstract.

We study the exit probability for several binary opinion dynamics models in one dimension in which the opinion state (represented by ± 1) of an agent is determined by dynamical rules dependent on the size of its neighbouring domains. In all these models, we find the exit probability behaves like a step function in the thermodynamic limit. In a finite system of size L , the exit probability $E(x)$ as a function of the initial fraction x of one type of opinion is given by $E(x) = f[(x - x_c)L^{1/\nu}]$ with a universal value of $\nu = 2.5 \pm 0.03$. The form of the scaling function is also universal: $f(y) = [\tanh(\lambda y + c) + 1]/2$, where λ is found to be dependent on the particular dynamics. The variation of λ against the parameters of the models is studied. c is non-zero only when the dynamical rule distinguishes between ± 1 states; comparison with theoretical estimates in this case shows very good agreement.

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1. Introduction

Identifying universality in physical phenomena occurring in different systems has become an important topic of research in the last few decades. Universality usually indicates that there are some fundamental common underlying features in the systems under consideration. It also signifies that there are only a few parameters occurring in the systems those are relevant. Existence of universality justifies the study of models which include only these parameters while real systems are far more complicated. Universality has been observed in critical phenomena; for example there is a unique value of the order parameter exponent in liquid-gas phase transition [1]. Universal features may also appear away from criticality as for example the universal scaling behaviour obtained for characteristic features of many complex networks [2]. In dynamical phenomena, universal classes have been observed close to and away from criticality. Models belonging to the same static universality class may belong to a different universality class as far as dynamics is concerned [3]. Non-equilibrium dynamics also reveal dynamical universal classes, for example, many systems have been shown to belong to the directed percolation universality class [4, 5].

Exit probability (EP) is one important feature of dynamical models with two absorbing states. Examples include binary opinion dynamics models and Ising spin models. Exit probability $E(x)$ denotes the probability that the system ends up with all opinion/spins in a certain state when initially x fraction had been in that state. Recently, a lot of effort has been put to identify universal features of exit probability in opinion dynamics models (as well as generalised Ising-Glauber models) in one dimension. In the voter model (which is equivalent to the Ising-Glauber dynamics in one-dimension) $E(x)$ is simply equal to x while $E(x)$ is a non-linear continuous function of x in nonlinear voter model, Sznajd model and long ranged Ising Glauber model [6, 7, 8, 9] in one dimension. In all these cases $E(x)$ apparently shows no dependence on finite sizes for larger system sizes [6]. The exit probability was also calculated and generalised for nonlinear q-voter model in one-dimension [10, 11, 12]. In higher dimensions or on networks, the exit probability in the thermodynamic limit may exhibit a step function behaviour, interpreted as a phase transition in some earlier works [13, 14, 15, 16, 17]. Strong finite size effects are observed here. The possibility of problems arising while calculating the exit probability, due to the sole use of local update rules in dynamical systems, was addressed previously by Galam et al [18].

In a recent study of a binary opinion dynamics model [19], the behaviour of the exit probability was found to be quite different from the well studied models mentioned in the paragraph above; it exhibited a step function like behaviour in the thermodynamic limit *even in one dimension*. It also showed the existence of an exponent with a value independent of the model parameter. In the model considered in [19], the state (spin or opinion) of an agent was updated according to a rule dependent on the size of his/her two neighbouring domains.

In this paper, we investigate whether the step function behaviour of $E(x)$ in the

thermodynamic limit is a universal feature of models with dynamical rules which involve the sizes of neighbouring domains in one dimension. Careful study of a number of models indicates that indeed such a universal feature exists. Furthermore, universal scaling function and an exponent with model independent universal value are obtained. The non-universal quantities associated with the scaling function also show very interesting behaviour as a function of the model parameters.

In section II, the models are introduced and details of the simulation provided briefly. Results are presented in section III and discussions and summary in the last section.

2. The Models

We have considered a number of models which mimic opinion formation in a society where the opinions have values ± 1 . These states can be equivalently regarded as the states of Ising spins and the models may be interpreted as interacting spin models as well. Since it is convenient to talk in terms of spins we will use the term spin instead of opinion henceforth. It also becomes more meaningful as the models behave as familiar spin models in certain limits.

In all these models considered in one dimension, in the spin picture, the spins located on the domain boundaries are liable to flip, as in the case of zero temperature Ising model with Glauber dynamics. The spins which can undergo change have therefore two neighbouring domains of opposite spin states. In general, in the models considered in this work, the state of the spin at the domain boundary is determined by the state of the neighbouring domains and their sizes.

2.0.1. BS model In the first model introduced in this class by Biswas and Sen [20], the BS model hereafter, the state of the spin simply follows that of the larger neighbouring domain. Hence if d_{up} and d_{down} are the neighbouring domain sizes (with up and down spins respectively), the spin will be up if $d_{up} > d_{down}$ and down otherwise. In case $d_{up} = d_{down}$, the state is chosen to be ± 1 with equal probability. A spin sandwiched between domains of opposite sign is always flipped. The BS model, where the final configuration is all up or all down states, is different from the Ising model having different dynamical exponents with respect to domain growth and persistence [20].

2.0.2. BS model with cutoff In the BS model one can introduce a cutoff [21] on the size of the domain while calculating d_{up} and d_{down} . The cutoff is taken as $R = pL/2$ where L is the system size and p a parameter ranging from zero to 1. Now, with the introduction of this cut off parameter p , the definition of d_{up} and d_{down} are modified: $d_{up} = \min\{R, d_{up}\}$ and similarly $d_{down} = \min\{R, d_{down}\}$ while the same dynamical rule explained earlier applies. When p is infinitesimal, the results are identical to those of the nearest neighbour Ising model. For finite p , there is a crossover behaviour in time: initially there is a BS-like behaviour after which very few domain walls survive which

perform almost noninteracting motion for a long time before annihilating each other. $p = 1$ is of course equivalent to the BS model. Instead of a fixed value of the cutoff a random cutoff can also be considered.

2.0.3. The β model We have also studied a model where the dynamics depends on the size of the neighbouring domains stochastically with a noise like parameter β [22]. In this so called β model, the probability of a boundary spin to be up is taken as

$$P(up) \propto e^{\beta(d_{up}-d_{down})}, \quad (1)$$

and it is down with probability

$$P(down) \propto e^{\beta(d_{down}-d_{up})}. \quad (2)$$

The normalised probabilities are therefore $P(up) = \exp \beta \Delta / (\exp(\beta \Delta) + \exp(-\beta \Delta))$ and $P(down) = 1 - P(up)$, where $\Delta = (d_{up} - d_{down})$. $\beta = 0$ is equivalent to the Ising model and any finite value of β drives the system to the BS dynamical class.

2.0.4. The ϵ model The BS model was shown to be equivalent to a reaction diffusion model in one dimension where random walkers tend to walk towards their nearer neighbours and annihilate on meeting [23]. This reaction diffusion model can be generalised by assigning a probability ϵ to move towards the nearer neighbour. $\epsilon = 1$ corresponds to the BS model and $\epsilon = 0.5$, the model with unbiased walkers which mimics the coarsening dynamics in the Ising model. We call this model the ϵ model. In the equivalent spin model, the larger neighbouring domain will dictate the sign of the spin on the boundary with probability ϵ .

2.0.5. The $\rho - \mu$ model Another stochastic model involving two parameters has been conceived [20]. Here a quenched disorder is introduced in the BS model through a parameter ρ representing the probability that people are completely rigid and never change their opinion throughout the time evolution. The second parameter μ relaxes the rigidity criterion in an annealed manner. It was found that although with $\rho \neq 0$ and $\mu = 0$, no consensus state is reached, any nonzero value of μ enables the system to reach the all up/down states [20].

2.0.6. The weighted influence (WI) model In all the models described above, the up and down states are taken to be indistinguishable. A model in which the up and down domains have different weight factors has been considered recently [19] and in fact the exit probability was also evaluated as mentioned in the introduction. We apply the analysis used in this work to the results of [19] to reveal certain interesting features related to the exit probability. This model was termed as the weighted influence (WI) model, where an individual takes up opinion 1 with probability

$$P_1 = \frac{d_{up}}{d_{up} + \delta d_{down}}. \quad (3)$$

δ is the relative influencing ability of the two groups and can vary from zero to ∞ . Probability to take opinion value -1 is $P_{-1} = 1 - P_1$.

All the models discussed above have one common feature in their dynamical rule. For all these models the state of the randomly selected spin depends on the size of the neighbouring domains somehow. In spite of this similarity the intrinsic dynamical rule of the models are different. The first one is the BS model which has no disorder, no stochasticity in the dynamics and the state of the selected spin becomes just the same as that of the larger neighbouring domain. Also in the BS model with cutoff there is no intrinsic stochasticity and disorder but its late time dynamical behaviour is completely different which is Ising like. In the β model, thermal noise like disorder is introduced. Here the dynamics is stochastic but for any non-zero β the system belongs to the dynamical class of BS model. In the $\rho - \mu$ model disorder is introduced through ρ and μ . For $\mu \neq 0$, this model has the same dynamical behaviour as the BS model. In the ϵ model, with $\epsilon > 0.5$, dynamical behaviour is same as BS model whereas for $\epsilon = 0.5$ the dynamics is Ising like and for $\epsilon < 0.5$ a different behaviour has been observed previously. On the other hand the WI model, which incorporates stochasticity, is not equivalent to the BS model in any limit. It belongs to a different universality class as far as persistence behaviour is concerned.

System sizes ranging from $L = 200$ to $L = 50000$ were considered depending on the model; e.g., since the BS model does not involve any parameter, one could probe much larger sizes \ddagger . All the simulations are done for at least 2500 configurations for each system size. Random sequential updating is used in all the simulations.

3. Results

3.1. Symmetric models: Exit probability

We first discuss the results for the behaviour of the EP for the models with symmetry where up/down states have the same status. These results show the existence of an exponent with a model independent universal value.

3.1.1. BS model For the BS model we have studied the exit probability $E(x)$ for system sizes ranging from $L = 6000$ to $L = 50000$. The plot of $E(x)$ against initial fraction of up spins x shows that it is nonlinear having strong system size dependence and that the different curves intersect at a single point $x_c = 0.5 \pm 0.001$ (shown in the inset of Fig. 1). The curves become steeper as the system size is increased. The exit probability thus shows a step function behaviour in the thermodynamic limit.

Finite size scaling analysis can be made using the scaling form

$$E(x, L) = f \left[\frac{(x - x_c)}{x_c} L^{1/\nu} \right] \quad (4)$$

\ddagger In fact to get reliable results for the dynamic exponents in the BS model one needs to simulate systems with size at least $\mathcal{O}(10^4)$ [20, 21, 22, 23].

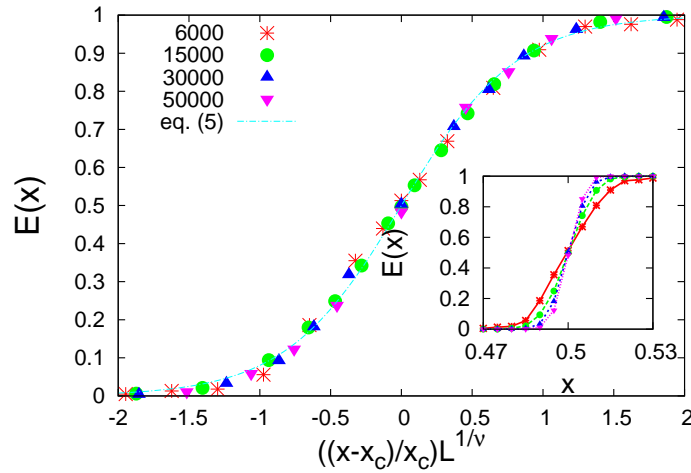


Figure 1. The data collapse of $E(x)$ plotted against $\frac{(x-x_c)}{x_c}L^{1/\nu}$ for different system sizes for BS model. Inset shows the unscaled data for the exit probability against initial concentration x .

where $f(y) \rightarrow 0$ for $y \ll 0$ and equal to 1 for $y \gg 0$, so that the data for different system sizes L collapse when $E(x)$ is plotted against $\frac{(x-x_c)}{x_c}L^{1/\nu}$. The data collapse takes place with $\nu = 2.5 \pm 0.03$ (Fig. 1). The value of x_c has been estimated as follows: ideally at x_c the exit probability is size independent. Numerically it is difficult to obtain exact intersection for all the system sizes. Above x_c , data for larger values of system size lie above in the $E(x)$ vs x plane, and below x_c the opposite happens, so by observing the range of x for which this happens, the error bars are estimated. To obtain ν , one uses the value of x_c obtained as above and the range of ν values for which the collapse appears to be good is taken to estimate the error bars. However, a good scaling collapse can only be obtained for sufficiently large values of L , typically $L > 5000$.

3.1.2. BS model with cutoff Introducing a cutoff in the BS model, we calculate $E(x)$ using cutoff factor $p > 0$ for different system sizes ranging from $L = 200$ to $L = 2000$. As long as $p < 1$, the time to reach equilibrium $\sim L^2$. So here we have to restrict system size at $L = 2000$ while for other models we use much larger system sizes.

Here $E(x)$ shows (Fig. 2) scaling behaviour given by eq. (4) as in the BS model with $\nu = 2.5 \pm 0.03$ and $x_c = 0.5 \pm 0.001$ for any p .

3.1.3. The β model In the β model we have studied the exit probability with noise parameter $\beta \geq 0$ for system sizes ranging from $L = 6000$ to $L = 30000$. For $\beta = 0$ the plot of $E(x)$ against x gives a straight line (Fig. 3) which is expected as it is identical to the nearest neighbour Ising model. In this case the result is independent of finite system sizes also.

For any nonzero value of β , EP shows nonlinear behaviour with strong system size

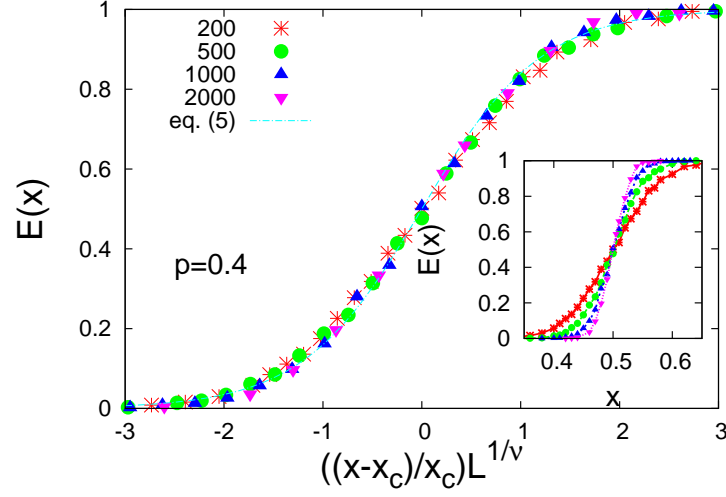


Figure 2. The data collapse of $E(x)$ plotted against $\frac{(x-x_c)}{x_c}L^{1/\nu}$ for different system sizes for BS model with cutoff $p = 0.4$. Inset shows the unscaled data for the exit probability against initial concentration x .

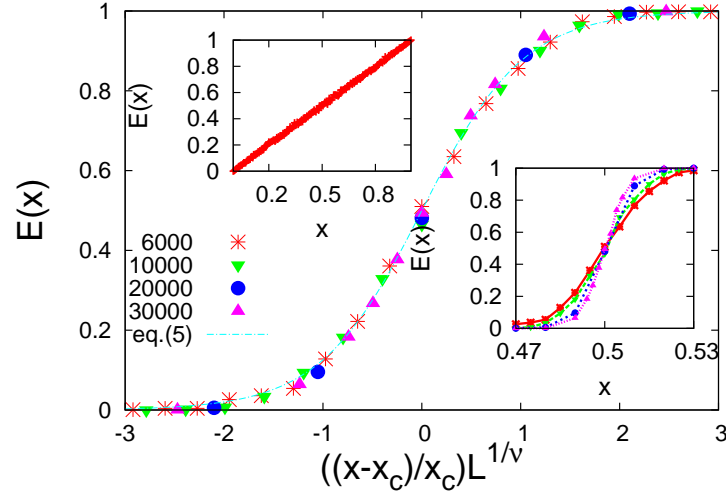


Figure 3. The data collapse of $E(x)$ plotted against $\frac{(x-x_c)}{x_c}L^{1/\nu}$ for different system sizes for model with noise parameter $\beta = 0.1$. Inset at the bottom shows the unscaled data for the exit probability against initial concentration x and inset at the top shows the plot of exit probability for $\beta = 0$.

dependence (Fig. 3) similar to the BS model. The scaling is once again found to be identical to eq. (4) with $\nu = 2.5 \pm 0.03$, independent of β . Here also $x_c = 0.5 \pm 0.001$ for all values of $\beta \neq 0$.

3.1.4. *The $\rho - \mu$ model* In the $\rho - \mu$ model we have studied the exit probability with rigidity parameter $\rho \leq 1$ for $0 < \mu \leq 1$ using different system sizes ranging from $L = 6000$ to $L = 30000$. For this model also, EP shows a nonlinear behaviour with strong system size dependence (Fig. 4). Here also $E(x)$ shows scaling behaviour given by eq. (4) with scaling exponent $\nu = 2.5 \pm 0.03$ independent of ρ and μ . Also $x_c = 0.5 \pm 0.001$ for all values of ρ and μ .

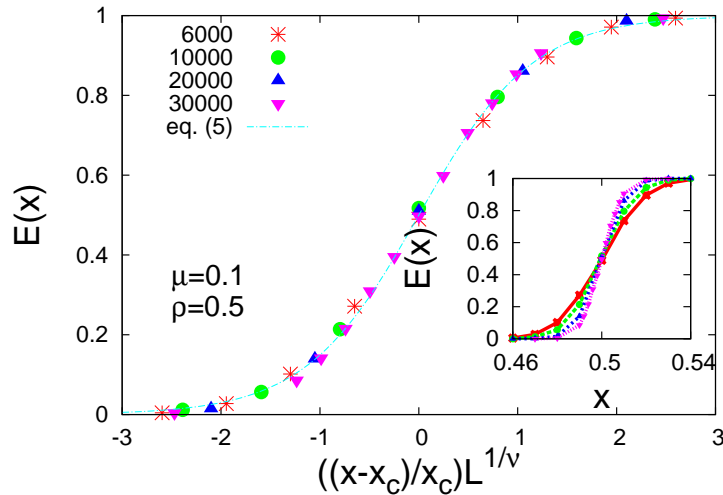


Figure 4. The data collapse of $E(x)$ plotted against $\frac{(x-x_c)}{x_c}L^{1/\nu}$ for $\rho - \mu$ model. Inset shows the unscaled data for the exit probability against initial concentration x .

3.1.5. *The ϵ model* For the ϵ model the system reaches the consensus state (all up/down) only for $\epsilon \geq 0.5$. In fact when $\epsilon < 0.5$, the final state is completely disordered.

$E(x)$ is simply equal to x for $\epsilon = 0.5$ which is expected as it is identical to the nearest neighbour Ising model. When $\epsilon > 0.5$, EP shows nonlinear behaviour with strong system size dependence (Fig. 5 shown for $\epsilon = 0.8$) for any value of ϵ .

The system sizes here vary from 5000 to 30000. Data collapse similar to the previously discussed models is also obtained here (Fig. 5) with $\nu = 2.5 \pm 0.03$ and $x_c = 0.5 \pm 0.001$.

3.2. Analysis of the scaling function for symmetric models

We find that in all the above cases, the EP becomes a step function at $x = 1/2$ in the thermodynamic limit and the scaling form given by eq. (4) is obeyed with the value of $\nu = 2.5 \pm 0.03$ being model independent. The scaling function f is found to fit very well with the general form

$$f(y) = [\tanh(\lambda y) + 1] / 2. \quad (5)$$

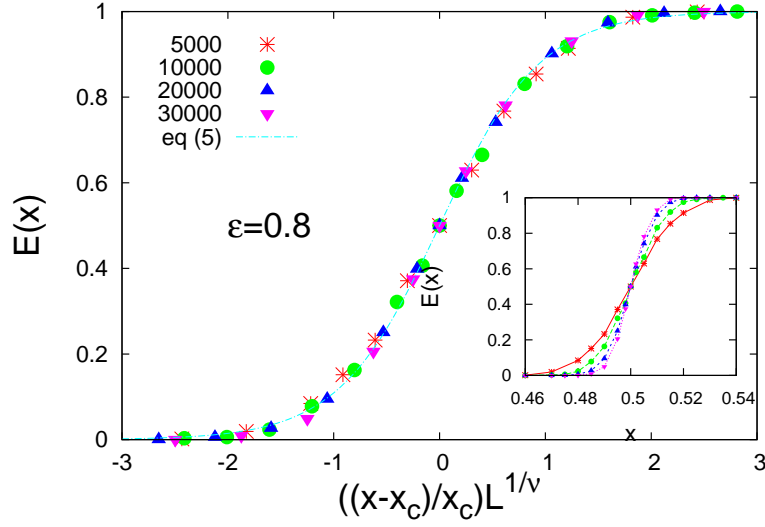


Figure 5. The data collapse of $E(x)$ plotted against $\frac{(x-x_c)}{x_c}L^{1/\nu}$ for different system sizes for ϵ model with $\epsilon = 0.8$. Inset shows the unscaled data for the exit probability against initial concentration x .

where $y = \frac{x-x_c}{x_c}L^{1/\nu}$. We conjecture the above form from the following considerations: first, the shape of the curve suggests a tanh form (note that the argument y varies from $-\infty$ to $+\infty$). Secondly, in the thermodynamic limit ($L \rightarrow \infty$), $E(x) \rightarrow 0$ for $x < x_c$ and $E(x) \rightarrow 1$ for $x > x_c$ such that one needs to add a factor of unity and also a division by 2 in $f(y)$. This form also leads to the result that $x_c = 1/2$ and $E(x_c) = 1/2$ in the symmetric models, shown later in this subsection.

We obtain the values of λ and find that λ is the factor which is different in each case (Figs 6, 7). For the BS model we found $\lambda = 1.22 \pm 0.02$ by fitting the collapsed plot of the model (fig 1) in equation 5.

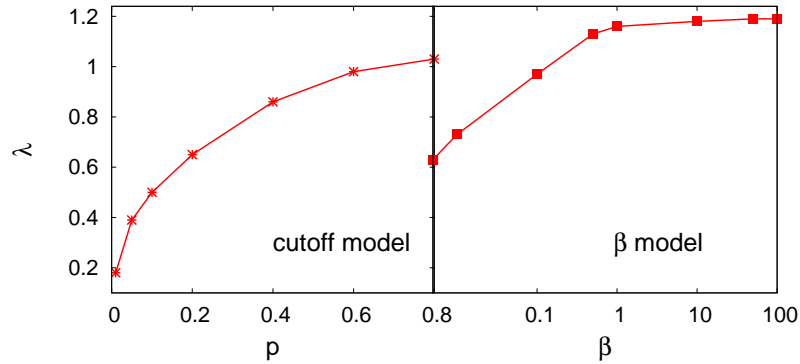


Figure 6. Plot of λ against cutoff p (cutoff model; left panel) and against β (noise model; right panel)

When the model involves a parameter, λ shows variation with the parameter value.

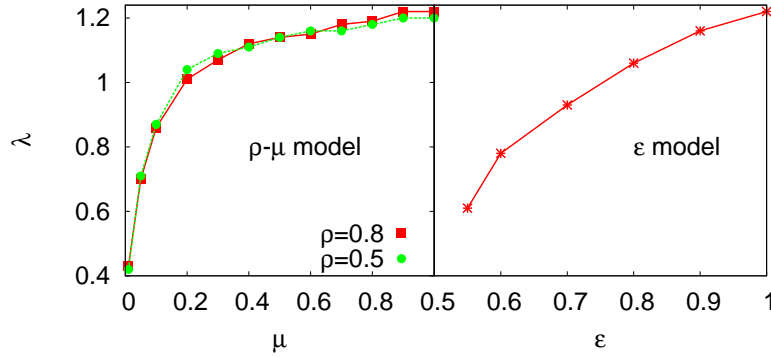


Figure 7. Plot of λ against parameter μ for two fixed values of $\rho = 0.5$ and $\rho = 0.8$ ($\rho - \mu$ model; left panel) and against parameter ϵ (ϵ model; right panel)

In the cutoff model, β model and the ϵ models, there are limiting values for which the models coincide with the Ising model and λ is undefined. On the other hand, for the $\rho - \mu$ model, for $\mu = 0$, λ is undefined as the dynamics do not lead to the all up/down state. As one increases the parameter values beyond that corresponding to the Ising limit in the first three models mentioned above and $\mu = 0$ in the $\rho - \mu$ model, one finds that λ increases. Each of these four models becomes equivalent to the BS model in the other extreme limiting values of the parameters used. Equivalence to the BS model is achieved in the cutoff model at $p = 1$; in the β model, for $\beta \rightarrow \infty$; in the ϵ model for $\epsilon = 1.0$ and in the $\rho - \mu$ model for $\mu = 1$ though the BS model behaviour may be present for even lesser values of the parameters as far as dynamical exponents are concerned. We find that λ varies monotonically and reaches a maximum value in the BS limit in general: $\lambda < 1.22$ in the cutoff model, ϵ model and $\rho - \mu$ model for other values of the parameter while in the β model, λ appears to assume the BS model value beyond a finite value $\beta \approx 1$.

In the cutoff model, a significant change in the timescale occurs as $p \rightarrow 1$ [21] and the behaviour of λ against p close to unity is no longer very smooth. For this reason, we show the results up to $p = 0.8$. The values of λ for the β model shows a rather intriguing behaviour: it has an increasing behaviour for $\beta < 1$ and beyond $\beta = 1$, λ increases very slowly and is almost a constant while approaching the BS value. However, there was no perceivable difference observed at $\beta = 1$ when other dynamical properties of this model were studied [22]. In the $\rho - \mu$ model, though λ depends on μ , we found it to be independent of ρ within error bars. It is due to the fact that ρ is an irrelevant parameter, while μ is a relevant parameter as shown in [20].

In general, one can now use eq. (5) to write down EP for the symmetric models as

$$E(x, L) = \frac{\left[\tanh \lambda \left(\frac{x - x_c}{x_c} \right) L^{1/\nu} + 1 \right]}{2}. \quad (6)$$

In all these models, $x_c = 1/2$ which can be established from symmetry arguments. In the BS model, one has no parameter and λ has a unique value. Using the relation

$$E(x) + E(1 - x) = 1, \quad (7)$$

and putting the expression for $E(x)$ from eq.(6), one can easily show that x_c has to be equal to $1/2$ and $E(x_c) = 1/2$. In the other models we find that λ has a dependence on the parameter value and in principle one can assume x_c to be a function of the parameter also, but the observed scaling form and the fact that eq. (7) has to be true for all x and L leads to the result that $x_c = 1/2$ always.

3.3. Analysis for the asymmetric WI model

In the WI model, it had already been noted that the value $\nu \simeq 2.5$ has to be used to obtain a data collapse. The EP here is found to be of the form

$$E(x, \delta) = \frac{[\tanh(\lambda(\delta)y + c(\delta)) + 1]}{2} \quad (8)$$

where $y = \frac{x-x_c}{x_c} L^{1/\nu}$; x_c, λ and c all vary with δ .

In this model, as the up and down spins have different status, it has to be noted that the probability that the final state is all down starting with x down spins will not be the same as $E(x, \delta)$ for up spins. Rather, to consider the negative spin case, one has to replace δ by $1/\delta$ such that

$$E(1-x, \delta) + E(x, 1/\delta) = 1. \quad (9)$$

We use the short-hand notation x_c for $x_c(\delta)$, c for $c(\delta)$ and λ for $\lambda(\delta)$. For $1/\delta$, we use primed variables, e.g. x'_c for $x_c(1/\delta)$. Putting the expression of $E(x, \delta)$ in eq. (9), we get

$$\tanh \left[\lambda \left(\frac{1-x-x_c}{x_c} \right) L^{1/\nu} + c \right] = -\tanh \left[\lambda' \left(\frac{x-x'_c}{x'_c} \right) L^{1/\nu} + c' \right].$$

On simplification one gets

$$c + c' = L^{1/\nu} \left[x \left(\frac{\lambda}{x_c} - \frac{\lambda'}{x'_c} \right) - \lambda \left(\frac{1-x_c}{x_c} \right) + \lambda' \right]. \quad (10)$$

Since $c(\delta)$ and $c(1/\delta)$ cannot have any L dependence then,

$$c(\delta) = -c(1/\delta). \quad (11)$$

For the right hand side of (10) to be zero for any value of x ,

$$\frac{\lambda(\delta)}{x_c(\delta)} = \frac{\lambda(1/\delta)}{x_c(1/\delta)} \quad (12)$$

and,

$$\lambda(\delta) \left[\frac{1-x_c(\delta)}{x_c(\delta)} \right] = \lambda(1/\delta). \quad (13)$$

Eliminating $\lambda(\delta)$ and $\lambda(1/\delta)$ from equations (12) and (13) we have,

$$x_c(\delta) = 1 - x_c(1/\delta) \quad (14)$$

which is consistent with the fact that x replaced by $(1-x)$ indicates δ replaced by $1/\delta$ which is the basis of eq. (9). We have checked that equations (12) and (13) show excellent matching with numerical data. For example for $\delta = 2.0$, we obtain

$\lambda = 1.6 \pm 0.01$ and $x_c = 0.696 \pm 0.001$ and (as $1/\delta = 1/2 = 0.5$) the corresponding values for $\delta = 0.5$ are $\lambda = 0.7 \pm 0.01$ and $x_c = 0.302 \pm 0.001$. Inserting these values we find nearly perfect agreement of both sides of equations (12) and (13). However, this can be checked for any other values of δ by extrapolating the curve shown in the bottom inset of Fig. 8 and indeed one can get good agreement.

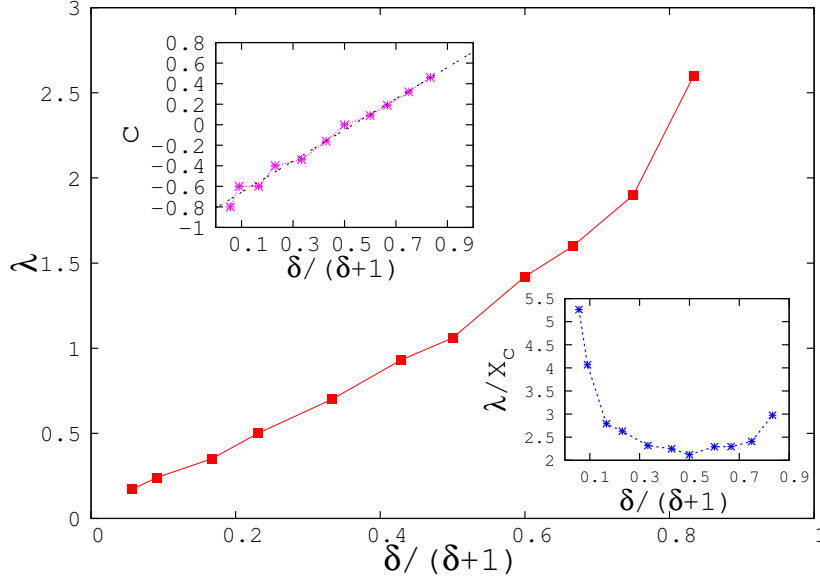


Figure 8. Plot of λ against parameter $\delta/(\delta+1)$ and the top inset shows variation of c with parameter $\delta/(\delta+1)$ and the bottom inset shows variation of $\frac{\lambda}{x_c}$ against $\delta/(\delta+1)$. The dashed straight line at the top inset is the fitted curve based on eq. (15).

Since the data for $c(\delta)$ has a lot of fluctuations for small values of δ (see inset of Fig. 8), a direct check of eq. (11) is difficult. However, from the figure $c(\delta)$ appears to vary linearly with $\delta/(\delta+1)$ and one can assume the form

$$c(\delta) = a \frac{\delta}{1 + \delta} - b. \quad (15)$$

In case eq. (11) is correct, one must have $b = a/2$. We tried the above form and the fit appears to be quite accurate with $a = 1.53 \pm 0.05$ and $b = 0.81 \pm 0.03$ which shows that b is very close to $a/2$ within error bar. This shows that indeed the scaling form we assumed is consistent with the theory.

In the case of WI model, where x_c is different for different δ , slope of $E(x)$ is determined by $\frac{\lambda}{x_c}$ which shows a minimum at $\delta = 1$ and increases otherwise (Fig. 8, bottom inset). Thus any asymmetry makes the EP steeper which is also expected as the asymmetry makes the system biased towards one of the absorbing states.

4. Summary and discussion

We have studied a number of opinion dynamics models in one dimension with different evolutionary rules for the state of the opinions/spins. A common feature is that the

update rules involve the size of the domains neighbouring a spin. This immediately gives rise to a different behaviour of the exit probability compared to that in the well-studied models in one dimension. It shows finite size dependence and a step function like behaviour in the thermodynamic limit. The step function occurs at $x = 0.5$ for models in which up and down states carry equal weight.

A scaling function with a universal form is found to exist with a universal value of the exponent ν occurring in it. Though the scaling function involves a tanh term is a conjecture, however, we have shown that such a conjecture leads to consistent and meaningful results. Two non-universal parameters λ and c appear in the scaling function. c is zero for models which have up/down symmetry. λ has strong model dependence and it shows interesting variation with the model parameters.

The scaling argument for the finite size scaling is $\lambda(\frac{x-x_c}{x_c})L^{1/\nu}$, which indicates that the width w of the region where $E(x)$ is not equal to unity or zero decreases as $\frac{x_c}{\lambda}L^{-1/\nu}$. When the rule is simply a majority rule, i.e., the larger neighbouring domain dictates the updated state of the spin (BS model), the value of λ is obtained as ≈ 1.22 . For the symmetric models which involve a parameter, λ is smaller than the BS value while x_c is still equal to 0.5. This signifies that w is larger for all these models compared to the BS model. In the WI model where up/down spins are distinguished, λ/x_c is larger than the BS value as δ deviates from unity showing that w in this case is smaller than the BS value. Asymmetry thus plays a strong role in determining the width w .

In the analysis of the WI model, one can further derive equations connecting the values of x_c , c and λ at δ and $1/\delta$ using eq. (8), which shows very good agreement with numerical data.

We thus arrive at the conclusion that there exists a class of models in one dimension that shows a behaviour different from familiar short range spin models in term of EP. Studying different models all of which use a dynamical rule involving the size of the neighbouring domains, a universal scaling behaviour accompanied by an exponent with universal value is obtained. The coarsening behaviour of the models considered here are not identical, e.g., the cutoff model has a Ising-like late time dynamics (domain growth exponent $z = 2$ [21]) while the other models show BS like behaviour ($z = 1$). Hence the step function behaviour of EP is clearly due to the domain size dependent dynamics as it is known that for the Ising model EP is just a linear function independent of system size. Asymmetry plays an important role but the value of ν is not affected.

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